

Solving Recurrence Relations



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Analyze

Some Functions are Easy

```
for(int i = 0; i < n; i++) {
```

$O(n)$

```
    for(int j = 0; j < n; j++) {
```

$O(n^2)$

```
        do_something_simple();
```

No early termination? Done!

```
    }
```

```
}
```

Sometimes We Could Be More Precise

```
for(int i = 0; i < n; i++) {
```

```
    for(int j = 0; j <= i; j++) {
```

$n(n+1)/2$, still $O(n^2)$

```
        do_something_simple();
```

Drop one "=", becomes $n(n-1)/2$, still $O(n^2)$

```
    }
```

```
}
```

Recursive Functions Can Be Tricky

```
int pow(int base, int ex) {
```

```
// only works on powers of 2
```

```
    if (ex == 0) return 1;
```

Constant time

```
    if (ex == 1) return base;
```

Constant time

```
    return pow(base, ex/2) * pow(base, ex/2);
```

Function of ex

```
}
```

Recursive Functions Can Be Tricky

```
int pow(int base, int ex) {
```

```
// only works on powers of 2
```

```
    if (ex == 0) return 1;
```

```
    if (ex == 1) return base;
```

```
    return pow(base, ex/2) * pow(base, ex/2);
```

```
}
```

$T(0) = C$

$T(1) = C$

$T(N) = 2 * T(N/2) + C_2$

Assumption that return/mult is constant

Substitute, look for pattern

$$T(N) = 2 * T(N/2) + C_2$$

$$T(N) = 4 * T(N/4) + 2 * C_2$$

$$T(N) = 8 * T(N/8) + 4 * C_2$$

$$T(N) = 2^{k*} T(N/2^k) + 2^{k-1} * C_2$$

Choose k such that $N/2^k = 1$

$$2^k = N, k = \lg N$$

$$T(N) = 2^{\lg N} * T(1) + 2^{\lg N - 1} * C_2$$

$$T(N/2) = 2 * T(N/4) + C_2$$

$$T(N/4) = 2 * T(N/8) + C_2$$

$$T(N/8) = 2 * T(N/16) + C_2$$

...

$$2^{\lg N - 1} = 0.5 * N$$

Substitute and Simplify

$$T(N) = 2^{\lg N} T(1) + 2^{\lg N - 1} C_2$$

$$T(N) = N * C + 0.5 * N * C_2$$

$$T(N) = O(N)$$

Master Theorem

Easy Memorization Way

$$T(n) = a * T(n/b) + f(n)$$

a is the number of sub-problems

b is the chunk size (assume equal size)

f(n) is the additional processing

You can memorize solutions for 3 special cases.

Aside, this is covered nicely:

https://en.wikipedia.org/wiki/Master_theorem

Following examples from Wikipedia

Case 1 - Divide and conquer dominates

If $f(n) = O(n^c)$ where $c < \log_b a$

$$T(n) = \Theta(n^{\log_b a})$$

$$T(n) = 8T(n/2) + 1000n^2$$

$$a = 8$$

$$b = 2$$

$$f(n) = 1000n^2$$

$$\lg 8 = 3 > c$$

$$T(n) = \Theta(n^3)$$

Case 2 - It all matters

If, for some $k \geq 0$, $f(n) = \Theta(n^c \lg^k n)$ where $c = \log_b a$

$$T(n) = \Theta(n^c \lg^{k+1} n)$$

$$T(n) = 2T(n/2) + 10n$$

$$a=2$$

$$b=2$$

$$f(n) = 10n$$

$$c=1 \text{ (} k=0 \text{)}$$

$$T(n) = \Theta(n \lg n)$$

Case 3 - Your other stuff dominates divide-and-conquer

If $f(n) = \Omega(n^c)$ where $c > \log_b a$

$$T(n) = \Theta(f(n))$$

$$T(n) = 2T(n/2) + n^2$$

$$a = 2$$

$$b = 2$$

$$f(n) = \Omega(n^2), c = 2, \log_2 2 = 1, 2 > 1$$

$$T(n) = \Theta(n^2)$$